

Day 1: Introduction to Machine Learning

ARISE 2020: ECE Machine Learning Lab

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Outline

- 1 Introductions
- 2 What is Machine Learning?
- 3 Course Outline
- 4 Setting Up Python
- 5 Problem Solving as an Engineer
- 6 Basics of Programming in Python
- 7 Matrices and Vectors
- 8 Visualizing Data
- 9 Lab: Plotting Functions
- 10 ML vs. AI, Why the Hype Today?
- 11 Supervised Learning

Tell us about yourself

- Write down the following pieces of information in your book:
 - Name
 - Grade
 - Where are you from?
 - What time did you wake up this morning (exact times please!)?
 - What is your favorite thing to do during Quarantine?
 - What do you want to get out of this class?
 - One rule you'd like to propose for the classroom
- Discuss what you've written down
- We'll visualize this data in python later today
 - [Link to excel sheet here](#)

Outline

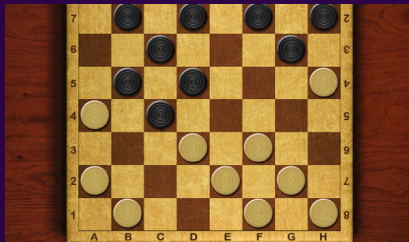
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Machine Learning

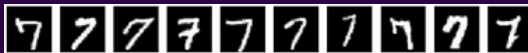
- Most recent exciting technology
- We use these algorithms dozens of times a day
 - Web search engine
 - Face detection
- Machine Learning is an important component to achieve the big AI dream
- Practice and critical thinking are the key to learn machine learning

Definition

- Machine Learning is a field of study that gives computers the ability to learn without being explicitly programmed.



Example: Digit Recognition



- Challenges with expert approach
 - Simple expert rule breaks down in practice
 - Difficult to translate our knowledge into code
- Machine Learning approach
 - Learned systems do very well on image recognition problems

```
def classify(image):  
    ...  
    nv = count_vert_lines(image)  
    nh = count_horiz_lines(image)  
    ...  
  
    if (nv == 1) and (nh == 1):  
        digit = 7  
    ...  
  
    return digit
```

Machine Learning Problem Pipeline

- 1 Gather data
- 2 Visualize the data
- 3 Formulate ML problem
 - Regression vs Classification
 - Choose an appropriate cost function
- 4 Design the model and train to find the optimal parameters of the model
 - Prepare a design matrix
 - Perform feature engineering
 - Validate your choice of hyper-parameters using a cross-validation set
- 5 Evaluate the model on a test set
 - If the performance is not satisfactory, go back to step 4

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Course Outline

- What is Machine Learning and why is it so relevant today?
- Linear Regression
- Logistic Regression
- Back propagation algorithm
- Mini-Project on topics covered so far
- Neural Networks
- Conv Nets
- PyTorch tutorial
- Applications of CNN's and RNNs
- Self-Driving Cars Final Project

Course Format, Website, Resources

- Course Website: github.com/ApoorvaNandini/ARISE2020
 - Github: share collections of documents, repositories of code
 - Contains lecture slides, code notebooks, and datasets
 - Slides posted before lecture, demo code and solutions posted after
- Strongly encourage programming in Python via Google Colab
 - No installation required
- Will give additional resources based on student interest

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Setting Up Python

- Scripting Python
 - Sequentially execute lines of code
- Jupyter Notebook
 - Code Blocks and Text Blocks (markdown)
 - Variables saved in workspace (across code blocks)
- Google Colab
 - Jupyter Notebook online (no real-time collaboration)
 - No installation
 - Free GPU for 12 hours
- Set up 1 of these 3 things and `print('hello world!')`

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Black-Box Abstraction

[i+-i]

- Can deal with many things with only knowing their **inputs** and **outputs**
- Need not know the details of the insides of our “boxes”
- Examples:
 - MTA turnstile
 - Python's `print()` function
 - Your Smart-Phone
 - Others?
- When does Black-Box Abstraction fail?

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Python Basics

- Program
 - We write operations to be executed on variables
- Variables
 - Referencing and interacting with items in the program
- If-Statements
 - Conditionally execute lines of code
- Functions
 - Reuse lines of code at any time

Python Basics

- Lists
 - Store an ordered collection of data
- Loops
 - Conditionally re-execute code
- Strings
 - Words and sentences are treated as lists of characters
- Classes (advanced)
 - Making your own data-type. Functions and variables made to be associated with it too.

Modules/Libraries/Packages

- NumPy: math, vectors and matrices
- Matplotlib: plotting graphs, visualizing data
- Pandas: convenient for storing and retrieving data
- Sklearn: convenient wrapper for simple ML problems
- TensorFlow: computational graph, neural networks
- Keras: convenient wrapper for TF
- PyTorch: deep learning framework, used mostly by Research community

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Vectors

- A **vector** is an n-tuple of numbers or symbols
 - Ex: integer 4-tuple: $\mathbf{v} = (1, 5, 2, 9)$
- Vectors of the same size may be added together, element-wise
 - Ex: $\mathbf{u} = (3, 6, 0, -5)$
 $\mathbf{v} + \mathbf{u} = (1 + 3, 5 + 6, 2 + 0, 9 + (-5)) = (4, 11, 2, 4)$
- Vectors may be scaled by a number, element-wise
 - Ex: $3\mathbf{v} = (3 \times 1, 3 \times 5, 3 \times 2, 3 \times 9) = (3, 15, 6, 27)$
- **Dot-Product**: sum of element-wise products of two vectors
 - Ex:
 $\mathbf{u} \cdot \mathbf{v} = 3 \times 1 + 6 \times 5 + 0 \times 2 + (-5) \times 9 = 3 + 30 - 45 = -12$
- Length of a Vector (L2 Norm) : $\|\mathbf{x}\| = \sqrt{\mathbf{x} \cdot \mathbf{x}}$
 - Ex: $\|\mathbf{v}\| = \sqrt{1 + 25 + 4 + 81} = \sqrt{111}$
- Dimension of a vector
 - Ex: $\mathbf{u} = (3, 6, 0, -5)$; $\dim(\mathbf{u}) = 4$

Matrices

- A **matrix** is a rectangular array of numbers or symbols arranged in rows and columns

- Ex: 2 by 3 matrix, $M = \begin{bmatrix} 1 & 9 & -12 \\ 15 & -2 & 0 \end{bmatrix}$

- Matrices of the same shape may be added together, element-wise

- Ex: $A = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix}, B = \begin{bmatrix} 0 & 8 \\ 7 & 11 \end{bmatrix}, A + B = \begin{bmatrix} 1 & 9 \\ 9 & 12 \end{bmatrix}$

- Matrices may be scaled, element-wise

- Ex: $aB = \begin{bmatrix} 0 & 8a \\ 7a & 11a \end{bmatrix}$, where a is a scalar

More About Matrices

- Matrices may be multiplied together provided their shapes meet the criteria
 - $(C)_{ij} = c_{ij} = \sum_{k=1}^N a_{ik} b_{kj}$
 - Criteria: for AB to be valid, # cols of A must equal the # rows of B
 - Result is a matrix with shape (# rows A , # cols B)
 - Ex: A is (2×3) and B is (3×5) , $C = AB$ is (2×5)
 - Example on board
- **Transpose:** A^T swaps the rows and columns of matrix A
- **Inverse:** A^{-1} satisfies the equation $AA^{-1} = A^{-1}A = \mathbb{I}$
 - Square matrices only!

Vectors and Matrices

- We may consider a vector as a matrix

- **Row Vector**: shape $(1 \times N)$

Ex: $\mathbf{v} = [1 \ 5 \ 2 \ 9]$

- **Column Vector**: shape $(N \times 1)$

Ex: $\mathbf{v} = \begin{bmatrix} 1 \\ 5 \\ 2 \\ 9 \end{bmatrix} = [1 \ 5 \ 2 \ 9]^T$

- We'll consider vectors as column vectors by default
 - Vector-Matrix Multiplication: on board
 - In what “dimension” is our vector starting, and where is it going to?
 - Dot Product:

Vectors and Matrices

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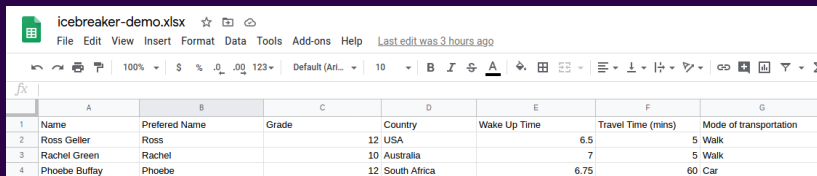
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Looking at our ice-breaker data in spreadsheets

- Columns have labels in the first row
- Collected data (numbers, words) follow below
- Let's export it to a Comma-Separated Values (CSV) file and open it



	A	B	C	D	E	F	G
1	Name	Preferred Name	Grade	Country	Wake Up Time	Travel Time (mins)	Mode of transportation
2	Ross Geller	Ross	12	USA	6.5	5	Walk
3	Rachel Green	Rachel	10	Australia	7	5	Walk
4	Phoebe Buffay	Phoebe	12	South Africa	6.75	60	Car

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Lab: Plotting Functions

- Generate and plot the following functions in Python:
 - Scatter plot of points: (0,1), (2,3), (5,2), (4,1)
 - Straight Line: $y = mx + b$
 - Sine-wave $y = \sin(x)$
 - Polynomial e.g. $y = x^3 + 2$
 - Exponential e.g. $y = e^{-2x}$
 - Gaussian (Use $\sigma = 0.5$)
 - Choose a function of your own
- Create separate plots for each of the functions, Compute the mean and variance of each function
- Use Wikipedia and Numpy Documentation to search for mathematical formulas and python functions

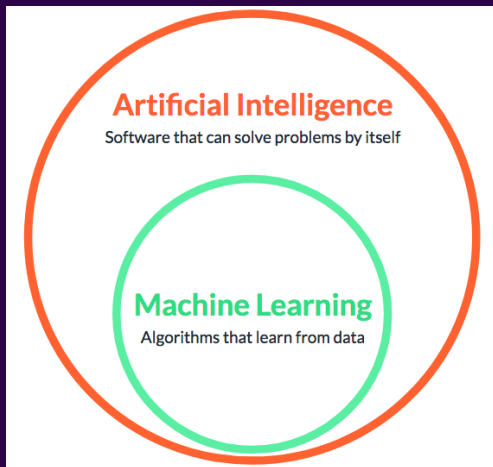
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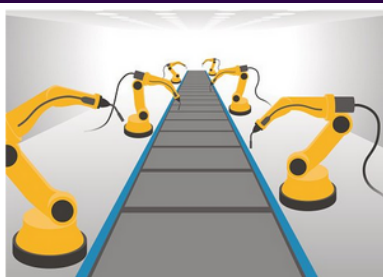
Artificial Intelligence

- Search
- Reasoning and Problem Solving
- Knowledge Representation
- Planning
- Learning
- Perception
- Natural Language Processing
- Motion and Manipulation
- Social and General Intelligence

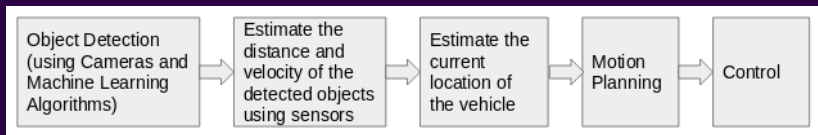
Machine Learning



Autonomous vs. Automated



Autonomous Example: Driver-less Cars



Why is Machine Learning so Prevalent?

- Database mining
- Medical records
- Computational biology
- Engineering
- Recommendation systems
- Understanding the human brain

Why Now?

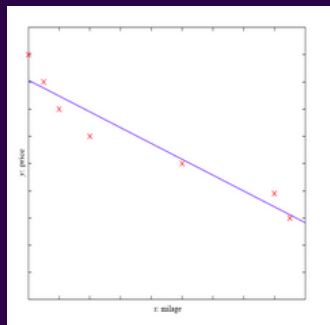
- Big Data
 - Massive storage. Large data centers
 - Massive connectivity
 - Sources of data from internet and elsewhere
- Computational advances
 - Distributed machines, clusters
 - GPUs and hardware

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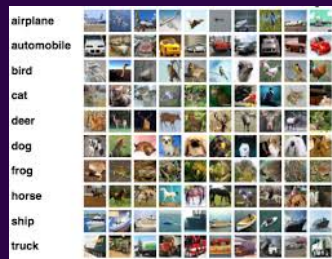
What is Regression?

- Target variable is continuous-valued
- Example
 - Predict $y = \text{price of a car}$
 - From $x = \text{mileage, size, horsepower}$
 - Can use multiple predictors
- Assume some form of mapping
 - Ex: Linear mapping: $y = b + w_1x$
 - Find parameter b, w_1 from data
- Use target-feature pairings as examples to form model



What is Classification?

- Determine what class a target belongs to based on its features
- Example:
 - Predict y = what type of object is in a photo
 - From x = the pixels of the image
- Learn a model/function from features to target
- Use target-feature pairings as examples to form model



Regression or Classification?

- **Problem 1:** Categorizing credit card applications into those who have good credit, bad credit and those who fall in the gray area.
 - Classification Problem
 - Class Labels: Good credit, Bad credit, Average credit
- **Problem 2:** Determining the sentiment of customer reviews for a product on Amazon.
 - Classification Problem
 - Class Labels: Positive, Negative, Neutral
- **Problem 3:** Predict whether an employee's income is over 100k a year or not.
 - Classification Problem
 - Class Labels: Over 100k, Under 100k

Regression or Classification?

- **Problem 4:** Estimating change in climate.
 - Regression Problem
 - Target Variable: Predicting future temperatures
- **Problem 5:** Identifying hate speech in social media.
 - Classification Problem
 - Class labels: Normal Speech, Hate Speech
- **Problem 6:** Forecasting the energy demand in a region.
 - Regression Problem
 - Target Variable: Predicting the amount of energy needed in the future

Thank You!

■ Next Class: ...